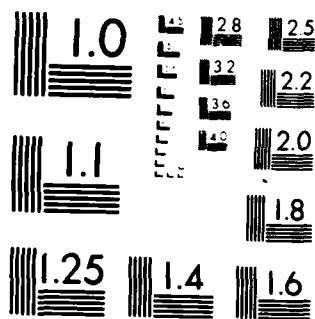


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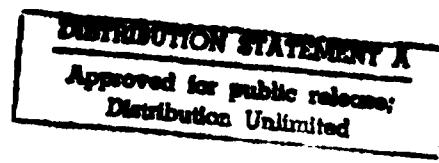
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Edward M. Connally

ACCURACY AND
COMPLEXITY
OF PROBLEM
SOLUTIONS WITH
EXAMINER
INTERACTIONS

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the specification of a logic for the formation of a novel task-based logic. The performance both of programmers and non-programmers increased with increasing levels of problem-complexity and with increasing computer processor support. For both age groups, errors-of-commission were relatively infrequent compared to errors-of-omission. It was found that the degree of processor-complexity was much more influential than problem-complexity in predicting performance score. The computer generalization of user-input was observed, performance was significantly lower than during all other experimental conditions. Results also showed that participant's training in the generation of problem solutions was a significant factor in performance, i.e., years-of-experience and years-of-education were not found to be good predictors of performance. The "cues-back-aides" were shown to be most effective when they included the logic implied by the example-solutions. These experiments demonstrate the effectiveness of the on-line use of computer software to create and modify software routines.

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ABSTRACT

This research evaluated the ability of computer users, both programmers and non-programmers, to specify problem solutions in the form of example-solutions. This ability was evaluated as a function of the complexity of the processor, i.e., the degree of generalization of the user input, the complexity of the problem, and the complexity of the feedback-aids. The experimental task employed in this study required the specification of a logic for the formation of a naval task-force. The performance both of programmers and non-programmers decreased with increasing levels of problem-complexity and with reduced processor-support. For both the groups, errors-of-commission were relatively infrequent compared to errors-of-omission. It was found that the degree of processor-complexity was much more influential than problem-complexity in predicting performance scores. When little computer generalization of user-input was provided, performance was significantly lower than during all other experimental conditions. Results also showed that participant-strategy in the generation of problem solutions was a significant factor in performance, though years-of-experience and years-of-education were not found to be good predictors of performance. The feedback-aids were shown to be most effective when they included the logic implied by the example-solutions. These experiments demonstrate the effectiveness of the on-line use of computer software to create and modify software routines.

INTRODUCTION

This is the final report on Contract N00014-79-C-0730, Work Order Number N8511-101 between Performance Measurement Associates, Inc., and the Engineering Psychology Group, Office of Naval Research. The contract was initiated on 1 Sept 1979 and ended on 28 Feb 1982.

The research investigated the capability of programmers and non-programmers to specify problem solutions by developing example-solutions and aids by writing computer programs; each method of specification was accomplished at various levels of problem-complexity. The level of difficulty of each problem was reflected by the number of steps needed by the user to develop a solution. Machine processing of the user-inputs permitted inference to be developed about the algorithms required to solve a particular problem. The interactive feedback of processing enabled the users to a more precise definition of the desired solution.

Six experiments were conducted, with the same problems used in all experiments. The ability of the participants to develop example-solutions was evaluated as a function of the participant's background and experience, the complexity of the problem to be solved, the level of processing provided by the computer, and the level of feedback-aids, when aids were available.

Two technical reports were published and three papers written, some documenting the experiments and results obtained. The experiments are identified in the list of technical reports and citations to the end of this report.

Experiments 1 and 2 were designed to investigate the ability of expert programmers and of bookkeepers/accountants who were not expert programmers to develop example-solutions for a hypothetical Navy task-force problem. The experimental variables in both experiments were problem-complexity and problem-on-complexity, i.e., the amount of machine processing of a user input.

Experiments 3 and 4 were designed to investigate the ability of expert programmers and non-programmers to develop accurate and complete example-solutions using various feedback-

aids at various levels of problem-complexity. The following aids designs were based on the results of six computer programs, where the systematic generation of example-solutions, a measure involving a combinational-measure, had been shown to correlate significantly with performance (explaining 70% of variance in classification).

Experiment 5 was designed to involve the use of a panel of expert programmers to revise and update the aids developed in the form of example-solutions to various ship combinations. One incorrect entries had been introduced in each of the aids developed in Experiments 3 and 4.

Finally, Experiment 6 utilized the computer system used to develop computer code written in the syntax of FORTRAN, at three levels of data input - a design interface, a reduced version of the design of Experiment 1. The results of Experiment 6 will be sub-routines written in FORTRAN IV that will calculate a correct ship combination, as that combination will be noted as uncorrected.

The performance measures used in the experiments above consisted of error-measures and strategies-measures. Three error-measures were:

- a. P_T , the probability that a given ship-combination was correctly classified as acceptable or unacceptable.
- b. P_C , the probability that a correct ship-combination was accepted.
- c. P_{IC} , the probability that an incorrect ship-combination was rejected.

In addition to the error-measures above, relative error-measures were used. A relative error-measure was defined as a participant's error-score (P_T , P_C , P_{IC}) on an experimental problem minus his/her error-score on the pre-test problem. The relative error-measures thus tended to remove the effect of the participant's innate capability, and, as a result, were more sensitive to experiment factors than were the error-measures alone.

Two strategy-measures were also introduced. One strategy-measure, the combinational-measure, detected the following situation: a participant changed only one component at a time of an example-solution. Another strategy-measure, the sequential-measure, detected the use-patterns of the various feedback-aids.

RESULTS OF EXPERIMENT 1

Experiment 1 and 2

The results of Experiments 1 and 2, according to the two categories, are described below. The design of Experiment 1 through 6 was largely based on 1971.

Processor Complexity and Example-Solutions

First, as expected, more errors occurred during initial work on the more complex problems. However, the level of processing, or generalization, of the example-solutions was found to be an important error-reducing factor, i.e., a significant reduction in errors occurred when data from example-solutions were processed into a standard form and presented to the participant.

Systematic Strategies and Feedback-Aids

A second result, and perhaps the most important, was that participants in both categories who performed well tended to use a systematic, step-by-step strategy in selecting example-solutions. This result, together with the first, noted above, suggested that feedback-aids might be designed to encourage participants to use systematic strategies, by processing their example-solutions and then feeding back the resultant data to suggest possible additional inputs.

Breadth vs. Depth of Experience

A third result of the first two experiments applied to the subsequent experiments was that the number of years advanced education (i.e., beyond high school) and the number of years

of professional experience were found to be uncorrelated in predicting performance. As a consequence of this finding, additional demographic factors were evaluated for their significance in the subsequent experiments, in an effort to identify more valid predictors of performance.

Low Frequency of Errors-of-Com-

The fourth result applied to the less-educated participants was the observation that only a few errors-of-commission were detected during the generation of the example-solutions. The majority of errors that did occur were errors-of-omission. This low error rate result influenced the design of Experiment 1, where FCRT-XANALY code was written to solve the same problem as did the less-educated, so that a comparison of errors-of-commission could be made.

Experiment 1

Feedback-Aids

Three aids were developed. Aid #1 provided the ship-selection logic (SSL) implied by the participants' example-solutions. Aid #2 included the SSL and an ordered listing of the ship-type of the example-solutions input by the participant. The ordered listing was intended to aid the participant by showing all possible omissions in ship-combinations. Aid #3 included the SSL and a list of suggested ship-combinations to consider. The suggested ship-combinations were those logically required to complete the combinations suggested by the example-solutions previously entered.

The effect of the feedback-aids as measured by the error-score (P_C), i.e. by the probability of accepting a correct ship-combination, was not statistically significant. However, the effect of the aids as measured by the relative error-score (error-score on the experiment problem minus the error-score on the pretest problem) was found to be significant and important. Apparently, the feedback-aids did help at least a portion of the participant population -- the less-than-superior performers. Those who would perform well without the aids were not helped by the aids. Also, it was apparent that variations in performance due to the participants' innate abilities were greater than variations in performance due to the feedback-aids. This factor,

plus the observation that supervisor performances may not necessarily use the aids, may account for the insignificant effect of the aids on error-score and the significant effect on relative-error-score.

Feedback-Aid #3 appeared to affect performance to a greater degree than Aid #2. Aid #3 included recommendations of next-logical example-solutions, whereas Aid #2 did not. Participants in the example-solutions input previously, Aid #2, on the other hand, included an ordered list of example-solutions previously, Aid #3. With Aid #2, the participant had to examine the patterns of previous inputs and identify any missing results or solutions. With Aid #3, the participant was given only one recommended solution. Given the example-solutions provided, it is conceivable that no errors would be reasonable to expect. Yet Aid #3 might support problem-solving better than that supported by Aid #2.

Aid #3 could give false sense of continuation., however, without indicating the aid. For the example-solutions input, for instance, Aid #3 was displayed in the form, where after a few example-solutions had been typed, the participant, if there were possible next-logical solutions, which were not complete. Also, when the participant inputted an example-solution that contained an error, Aid #3 would recommend solutions that did not complete the error-induced pattern. See aids as Aid #3 can thus be either helpful or harmful, depending on how they are used. When recommended solutions are used without careful evaluation, this type of feedback-aid is potentially harmful. If, however, the display of patterns built upon a error increases the likelihood of the error's detection - by displaying its impact - then aids that provide recommended solutions may help the user to detect such errors.

Breadth vs. Depth of Experience

The lack of a strong predictive relationship between years-of-higher-education or years-of-experience and performance may come as a surprise to educators and directors of personnel departments. This result was found in all of the experiments, so that very strong evidence is available to support the assertion that years-of-education and relevant work-experience are not good predictors of problem-solving performance.

Additional results suggest that the "number of programming languages (used on 1 or more occasions)" and "number of operating systems used" are better predictors of the characteristics of computer users/progammers.

Combinational Strategy

Combinational strategy was found to be the best predictor of performance. This result is reflected in Experiments 1 and 2, where variance explained after control variance was explained by combinational strategy (.61) in Experiments 1 and 2 and .67 in Experiment 3 (see Fig. 4). This indicates that the combinational-strategy measure had greater predictive power than moment-to-moment measures (combinational-moment and momentary). A moment-to-moment measure, however, was also found to be related to performance and thus provided possibly important diagnostic sensitivity.

Experiment 5

Feedback-Aids

In Experiment 5, participants were asked to revise example-solutions that included various numbers of errors and combinations. This experiment design permitted calculation of the probability of maintaining an initially-correct example-solution and the probability of detecting and correcting an initially-incorrect example-solution.

Analyses of grand-mean probabilities of success revealed that the probability of maintaining an initially-correct solution was .93. However, there was a probability of .65 for detecting and correcting an erroneous example-solution. Obviously, there was a performance decrement in the detection and correction of erroneous example-solutions. Useful feedback-aids might thus be directed toward the detection and correction of existing errors.

Feedback-Aids #2 and #3, which were the same as those used in Experiments 3 and 4, were not, however, a benefit to participants engaged in the revision of example-solutions. Further, an analysis using relative measures, which tended to remove the effect of participant skill, did not result in statistically significant results. The conclusion, then, was that the strategies used

successfully to reveal existing solutions. It is noted, however, that that the aids demanded to help in the development of new "clues" were not helpful in revealing incorrect and unused solutions.

Relativity of Error Detection

But there appears to have been a marked decrease in success in detection of errors. As the number of incorrect solutions found, the solutions were increasing in error-rate, a direct result of the difficulty of detecting and correcting the introduced example-solutions. The fewer the number of errors, the lower was the probability of detecting a given error.

This result suggests that there may be a base-line probability (based on the frequency of errors recently detected) of detecting and correcting errors, which affects the probability of judging that any solution is incorrect. Thus, the "correctness" of a solution as determined by a user may be a function of:

1. The actual correctness of that solution,
2. The perceived frequency of erroneous solutions found recently.

This hypothesis, consistent with predictions of signal detection theory (which says that the probability of an event is a function of a base-line probability in addition to specific measurements on the signal itself), predicts a decreasing probability of detecting and correcting errors with decreasing error-rates. This means that the probability of detecting the last few errors may be so small that it is not likely that they will be found. But it also suggests a possible solution: seeding errors (which can later be removed if not detected) to increase the base-line error rate and, thus, to increase the probability of detecting unknown errors.

Experiment 6

Two types of errors were analyzed. One type, termed an "error-of-omission", referred to an error that resulted in a failure to accept a correct entity (e.g. ship combination). When specifying a problem solution with example-solutions, an error-

of-omission could be avoided by having the subject enter an "entity" of a suitable entity (e.g., "oranges"). An example type of error considered was an "error-of-commission." Two example-solutions were used to measure accuracy. In this study, an error-of-omission corresponded to an error of commission, and an error-of-processing which was taken to mean an error of omission or commission. For example, An error-of-commission would be committed by accepting incorrect entries from the subject.

Errors-of-Omission

There was little effect of problem complexity on errors-of-omission. The mean rate of specifying problem solutions, i.e., errors-of-omission, was 1.6% for FORTIRAN IV subroutine.

Errors-of-Commission

When specifying problem solutions via computer keyboard, the rate of errors-of-commission was dependent on problem complexity-level (Experiment 1) and on program mode (Experiment 10 Metric).^{*} For, given a constant problem complexity, such as in Experiment 8, the errors-of-commission could be eliminated, as evidenced by the 82% solution rate which performance degradation was experienced.

The most important finding concerning errors-of-commission was that specification by example-solution was superior to specification by program code. Analysis of the mean scores from Experiments 1, 2, and 3 provided strong evidence that using example-solutions substantially reduced errors-of-commission when compared to using FORTIRAN IV program code. The 35% rate for errors-of-commission with example-solution, compared favorably with 18% for program code.

Three hypotheses concerning the superior performance of the example-solution method seem plausible:

1. It was working with examples and dealing with each individual combination of items one-at-a-time that resulted in a low rate of errors-of-commission.

*For a discussion of Halstead's L Metric, see Connelly, Comeau & Johnson, Technical Report 81-301, 1981.

2. It was the identification of errors in program code at-a-time. Solutions were given in pairs. For each pair, all computer programs were examined. If one of the two solutions contained an error, the rate of errors-of-commission would be halved.
3. The user could then examine each solution individually, in pairs, to determine which of the two solutions had errors. If either one, or both, of the two solutions contained errors, the user could select different forms, or combinations of forms, if it was felt that the user did not want to be forced to view the incorrect solution in any way other than reduced or viewing each solution individually. Consequently, if predictions were confirmed by the user, he were free from examining incorrect logic forms and fed back to the user for analysis, a low rate of errors-of-commission would be attained.

These hypotheses are not alternative hypotheses - all could be true. We have strong evidence that the first hypothesis is true. If the second is true but not the third, program-design and coding methods could be adapted to a more computation-dependent structure. And finally, if the latter hypothesis were found to be true, pre-computation may be best suited to convert the user's program code into another form (while maintaining the same program logic) for feedback to the user.

CONCLUSIONS

1. Feedback-aids, to support use of example-solution, should include the logic implied by the example-solutions as well as new example-solutions that complete the logic patterns suggested by the existing set of example-solutions.
2. Feedback-aids consisting of an ordered listing of all present solutions also reported high performance, but, the predictive aid type referred to above is preferred.

3. Feedback-aids assist in improving the performance of both programmers and non-programmers who cannot perform well without assistance.
4. Performance measures are used to detect significant performance improvement with feedback-aids. These may be relative measures, which indicate the difference between performance on a common task and on an experimental task.
5. The lack of a strong relationship between "years-of-higher-education", "years-of-experience", and performance, coupled with the strong relationship between "number of computer languages" known and "number of operating systems" used, suggests that education and experience should not be a factor they have been in the past for hiring, promotion, determining salary level, and assigning tasks. Instead, the number of computer languages known and the number of operating systems used, rather than used until the underlying factors involved in tasks are discovered.
6. Apparently, the depth of an individual's experience is not as important to performance as is the breadth of his experience.
7. A possible common underlying experience-related factor is the ability to view problems from alternative viewpoints, or the ability to develop alternative approaches to problems - an ability that might be enhanced with feedback-aids.
8. The performance-prediction capability of strategy-measures, developed as moment-to-moment measures, not only clearly demonstrates that systematic strategies were used by successful participants (which led to the design of the feedback aids), but also convincingly demonstrates that moment-to-moment measures provide the sensitivity to explain considerable performance variance (approximately 60% in Experiments 1 thru 4).

9. When modifying an initially-incorrect example-solution, the probability (.1%) of maintaining an initially-correct example-solution was approximately the same as the probability of developing a new correct example-solution, i.e., the joint probability of detection and correction of an erroneous example-solution was low (.1%). This suggests that new rules should be developed to assist in the detection and correction of existing errors.
10. The probability of detection and correction of an erroneous example-solution decreased as the initial number of erroneous example-solutions was decreased. The fewer the total number of errors, the less was the likelihood of detecting a given error. This suggests that a method for increasing the probability of detecting errors is to seed errors, unknown to the individual reviewing the example-solutions (or computer program) but otherwise recognized, to increase the on-line rate of error detection and therefore to increase the probability of detecting an unknown error.
11. Software error-categories are typically defined only to facilitate data collection and recording. However, analysis of software errors showed that when an error category was decomposed into sub-categories, the independent variables in the prediction equations changed. It is concluded, therefore, that software error-categories should be selected with regard to predictability as well as to data collectability. Two sub-categories should be combined only when their prediction equations are homologous, i.e. have the same independent variables.
12. The superior performance (fewer errors-of-commission) achieved when using example-solutions and inductive processing to specify problem solutions over the performance achieved when using FORTRAN IV code may provide a basis for determining the underlying mechanism for that success and a means for incorporating that

mechanism into problem-solving tasks more effectively. Apparently, superior performance was generated either because each combination of two input variables was treated individually and/or because the example-solutions were transformed into another form from the input selection to an output. If the former is a significant factor, then the example-solution should be adapted to problem-solving tasks by computer-aids. If the latter is a significant factor, then a strategy-and-equipment should be developed to transform the logic provided by the user into another form which is then fed back to the user for his review. Such a transformation might prevent the program's equivalent logic.

RECOMMENDATIONS FOR FURTHER RESEARCH

1. The strategy-measures used to analyze FORTRAN IV were not moment-to-moment measures. Instead, they were a classification of types of possible strategies. The predictive power of the measures was only moderate compared to those used to evaluate performance in developing example-solutions. It is suggested that moment-to-moment strategy-measures be developed for both program-design and program-code tasks.
2. Feedback-aids designed to support development of original example-solutions were not found applicable to the revision of erroneous example-solutions. Since use of example-solutions is a viable way to specify problem solutions in itself, and reveals ways of improving program design and code, successful revision-strategies should be identified, and revision-aids should be derived from those strategies.
3. The "number of programming languages" known and the "number of operating systems" used have been shown to be good predictors of performance both in developing example-solutions and in writing program code. It is also suggested that the ability to develop alternative approaches may be a common

factor which may be enhanced by learning new languages and operating systems, and which may be a key performance-factor. The underlying factor or factors resulting in superior performance should be determined to assist in performance definition and design of improvement aids.

4. It is suggested that error-detection capability may be a function of a base-line error-detection rate, and therefore that certain errors unknown to the individual checking the material may increase the probability of detecting an unexpected error. This conjecture should be explored. An example would test to determine if writing performance is improved, and, if so, what frequency and type of error analysis should be used.
5. The basis for superior performance with example-solutions needs to be resolved. The concept of writing program code for each combination of factors (or the use of an aid to automatically analyze logic to help develop accurate combination-independent logic) and the concept of code transformation into a different logical form for feedback to the user for approval need to be contrasted in an experimental environment. There is a potential here for substantially increasing the correctness of computer programs if an aid can be developed to permit transfer of the superior, almost error-free performance with example-solutions to code-writing performance.
6. Independent of the methods for improving performance in writing program code suggested in Recommendation 5, a new aid design should be considered that combines general statements written in program code with redundant example solutions -- i.e. with example-solutions that are not part of a program test, but, instead, that are inductively transformed into an alternative code. A pre-compiler aid would produce actual code from both sources. Potential performance-improvement could provide high-quality, almost error-free code.

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TECHNICAL ARTICLES

Connelly, E. M., Connolly, S. J., & Johnson, M. J. The effect of automatic problem power and feedback integration on the accuracy and completeness of computer program solutions. Paper presented at the Human Factors Society Annual Conference, Atlanta, Georgia, October 1981.

Connelly, E. M. A comparison of the accuracy and completeness of problem solution performance and computer program solutions as a function of computer program complexity. Paper presented at the Human Factors Measurement Associates, Boston, Massachusetts.

TECHNICAL ARTICLES

Connelly, E. M., & Johnson, M. J. The effect of computer power and increasing levels of automation on the accuracy and completeness of computer program solutions. Paper presented at the Human Factors Annual Conference, Seattle, Washington, October 1982. (Ad M. J. Connelly, University of Maryland, College Park, Maryland, 1982).

Connelly, E. M., & Johnson, M. J. The effect of computer power and problem complexity on the accuracy and completeness of computer program solutions for computer programs. Paper presented at the Human Factors Society Annual Conference, Rochester, New York, October 1981.

Connelly, E. M. The effect of power, complexity, computer power, and feedback integration on the accuracy and completeness of computer program solutions. Paper presented at the Human Factors Society 30th Annual Conference, Seattle, Washington, October 1986.

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